## PROOF APPENDIX

PROOF. In the following, we will prove that it is difficult and sometimes even impossible to set a proper confidence threshold to make the distinguish between  $\mathbf{M}_{all}$  and  $\mathbf{M}_{mix}$ .

First, based on the definitions, if  $R_{fp}$  and  $R_{fn}$  satisfy the uniform-random distribution, the confidence of root-cause nodes  $\mathbf{M}_{all}$  and their children nodes can be computed as

$$C_{all} = \frac{Leaf_{all} \times (1 - R_{fn})}{Leaf_{all}} = 1 - R_{fn} \tag{1}$$

The confidence of the first kind of non-root-cause nodes  $\mathbf{M}_{none}$  can be computed as

$$C_{none} = \frac{Leaf_{none} \times R_{fp}}{Leaf_{none}} = R_{fp}$$
 (2)

For the second kind of non-root-cause nodes  $\mathbf{M}_{mix}$ , the  $\gamma$  percentage of leaf nodes descended from  $\mathbf{M}_{mix}$  are descended from the root-cause nodes  $\mathbf{M}_{all}$ , while  $1-\gamma$  percentage of leaves descended from  $\mathbf{M}_{mix}$  are also descended from the non-root-cause nodes  $\mathbf{M}_{none}$ . For example, if  $\mathcal{M}_1 = \{d1, l1, *\}$  is the unique root-cause node  $\mathbf{M}_{all}$ ,  $\mathcal{M}_2 = \{d2, l1, *\}$  is a non-root-cause node  $\mathbf{M}_{none}$  and  $\mathcal{M}_3 = \{*, l1, *\}$  is a non-root-cause node  $\mathbf{M}_{mix}$ . Then the confidence of  $\mathbf{M}_{mix}$  can be computed as

$$\mathcal{C}_{mix} = \frac{Leaf_{mix} \times \gamma \times (1 - R_{fn}) + Leaf_{mix} \times (1 - \gamma) \times R_{fp}}{Leaf_{mix}}$$
$$= \gamma - \gamma \times R_{fn} + R_{fp} - \gamma \times R_{fp}$$

We use a confidence threshold  $\delta_2$  to filter the graph nodes with low *confidence* value. If an ideal anomaly detection with 100% accuracy is applied, the *confidence* is 100% for both the root-cause nodes  $\mathbf{M}_{all}$  and their subset, while the *confidence* is 0% for all other non-root-cause nodes  $\mathbf{M}_{none}$ . If  $\delta_2 > (1 - R_{fn})$ , then we will miss all the root-cause nodes  $\mathbf{M}_{all}$ . If  $\delta_2 < R_{fp}$ , then we will report the non-root-cause

nodes  $\mathbf{M}_{none}$  as root-cause nodes falsely. Then an effect confidence threshold  $\delta_2$  should satisfy  $R_{fp} < \delta_2 < 1 - R_{fn}$  and  $R_{fp} + R_{fn} < 1$ .

It is difficult to find an appropriate confidence threshold  $\delta_2$  to identify the non-root-cause nodes  $\mathbf{M}_{mix}$  and their children nodes correctly. It can be proved that if  $R_{fp} + R_{fn} < 1$  then  $\mathcal{C}_{none} < \mathcal{C}_{mix} < \mathcal{C}_{all}$  i.e.,  $R_{fp} < \mathcal{C}_{mix} < 1 - R_{fn}$ . We first prove that  $\mathcal{C}_{mix} < \mathcal{C}_{all}$  as follows:

$$\therefore R_{fp} + R_{fn} < 1 \iff R_{fp} < 1 - R_{fn} 
\therefore 1 - \gamma > 0 
\therefore (1 - \gamma) \times R_{fp} < (1 - \gamma) \times (1 - R_{fn}) 
\iff \mathcal{C}_{mix} < \mathcal{C}_{all}$$
(3)

Similarly,

$$r^{1} = \sqrt{(x^{1} - x)^{2} + (y^{1} - y)^{2} + (z^{1} - z)}$$
 (5)

$$r^{2} = \sqrt{(x^{2} - x)^{2} + (y^{2} - y)^{2} + (z^{2} - z)}$$
 (6)

$$r^{3} = \sqrt{(x^{3} - x)^{2} + (y^{3} - y)^{2} + (z^{3} - z)}$$
 (7)

With  $\gamma$  changing between (0,1) constantly,  $C_{mix}$  changes with  $(R_{fp}, 1-R_{fn})$  constantly. Since the  $R_{fp}$  and  $R_{fn}$  can not be a perfect uniform-random distribution in practice, there is no clear boundary between  $\mathbf{M}_{all}$  and  $\mathbf{M}_{mix}$ . This completes the proof.